



Behavioral Modeling for Science & Energy Policy

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**"When we try to pick out anything by itself, we find it
hitched to everything else in the Universe."**

John Muir, [My First Summer in the Sierra](#) (Boston: Houghton Mifflin, 1911)

Contents

- **Behavior in general**
- **Specific challenges for basic science**
- **Way forward**

Behavior in the Real World

- **Experts fail to forecast**
- **Markets boom and bust**
- **Policy instruments fail to achieve expectations**
- **Large firms fail with little warning**
- **Change is crisis-driven rather than controlled**
- **New technologies are implemented during depressions**
- **Organizational structures resist change: bosses, hierarchies, bureaucrats**

Behavior in Dynamic Decision Making Experiments

- As the dynamic complexity of a system grows (as there are more time delays, feedbacks (especially positive feedbacks), accumulations (stock and flow structures) and nonlinearities, human performance worsens, and learning slows
- Decision makers seldom plan or implement their plans
- Dynamic complexity induces market overshoot and oscillation (Kampmann 1992)
- Hill climbing is difficult in simple resource systems, leading to collapse even when property rights are explicit (Moxnes 1998)
- Mental models of climate change violate conservation of matter (Sterman & Booth Sweeney 2002)

Behavior in The Majority of Climate Policy Models

- **Equilibrium is assumed rather than emergent**
- **Market outcomes reflect agent preferences**
- **Agents perceive and respond to prices instantaneously, and may even know the future**
- **Agents have sufficient structural knowledge to respond appropriately to changes in their environment**
- **There are few if any externalities other than climate**
- **Risk is absent**

Slide 6

TF7

There has been progress in bottom-up elements of large scale models, but the realm of compact models remains neoclassical.

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Consequences of Neglecting Behavior

- **Part of the observed inflexibility of the energy-economy system is behavioral; if it is instead ascribed to technology,**
 - estimates of the welfare consequences of changes in allocations are biased
 - the potential for institutional and informational policies is understated
- **Revealed preferences are suspect**
 - utility functions that justify observed savings and income distribution are unfair to future generations and today's poor
- **Idealized policy instruments won't deliver in the real world**

TF8

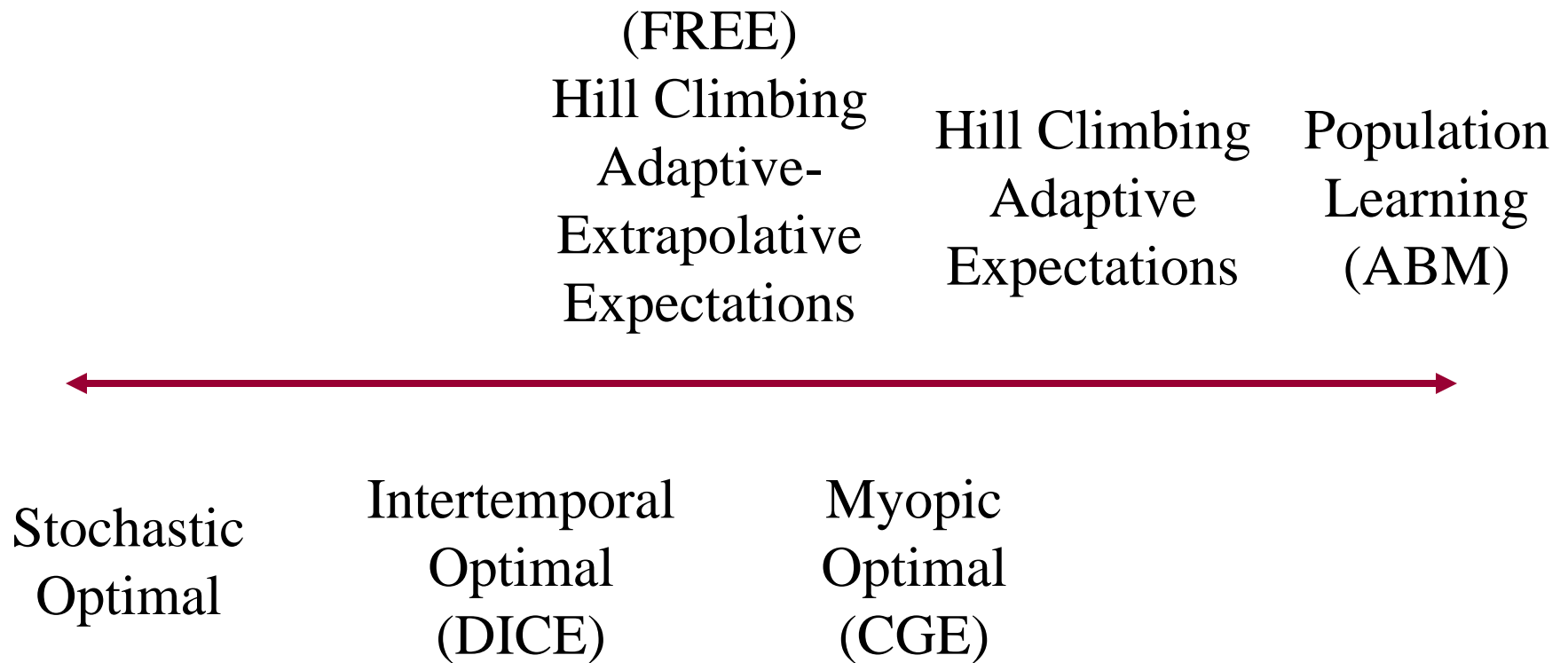
Slide 7

TF8

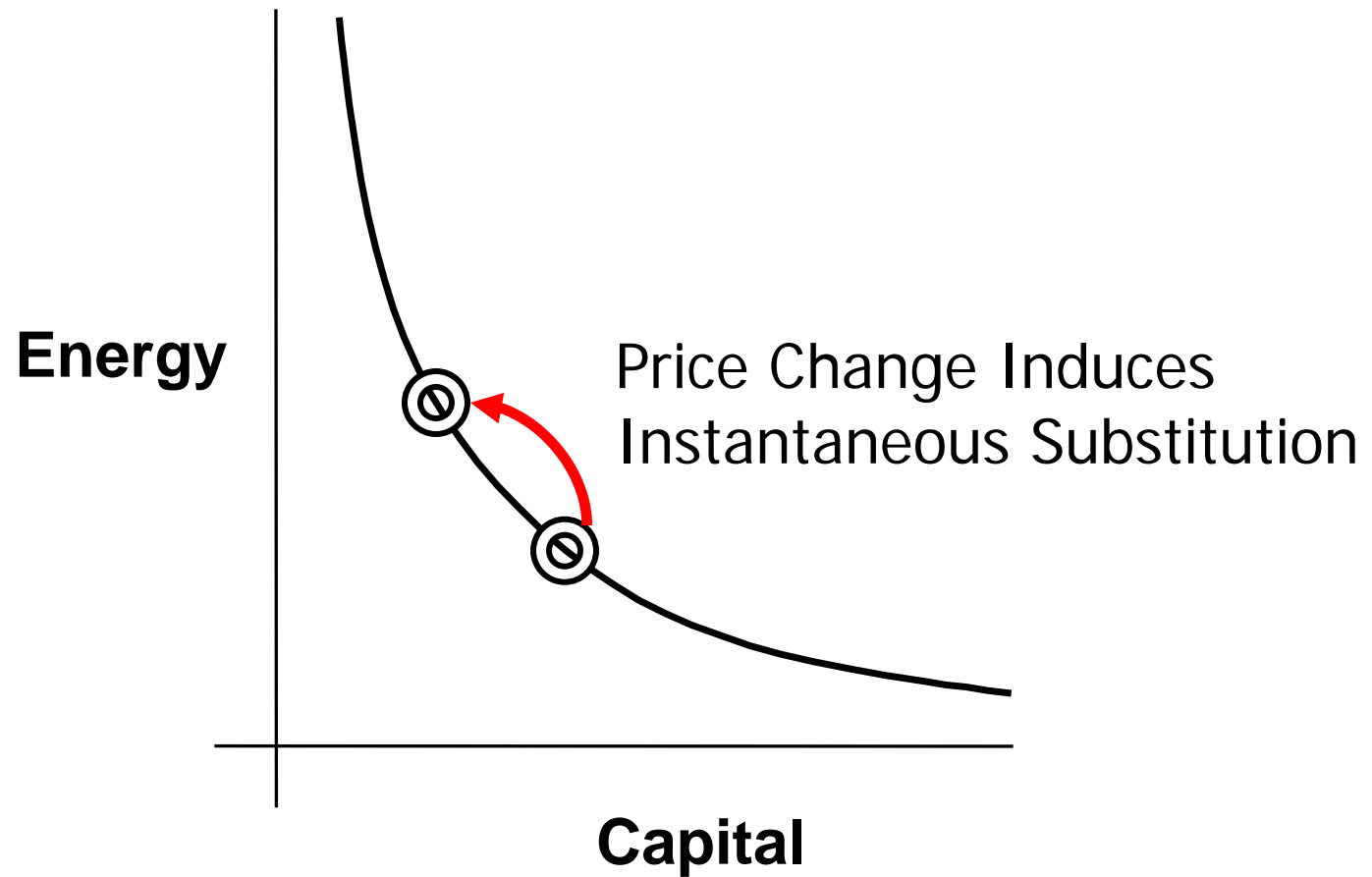
Lock in?

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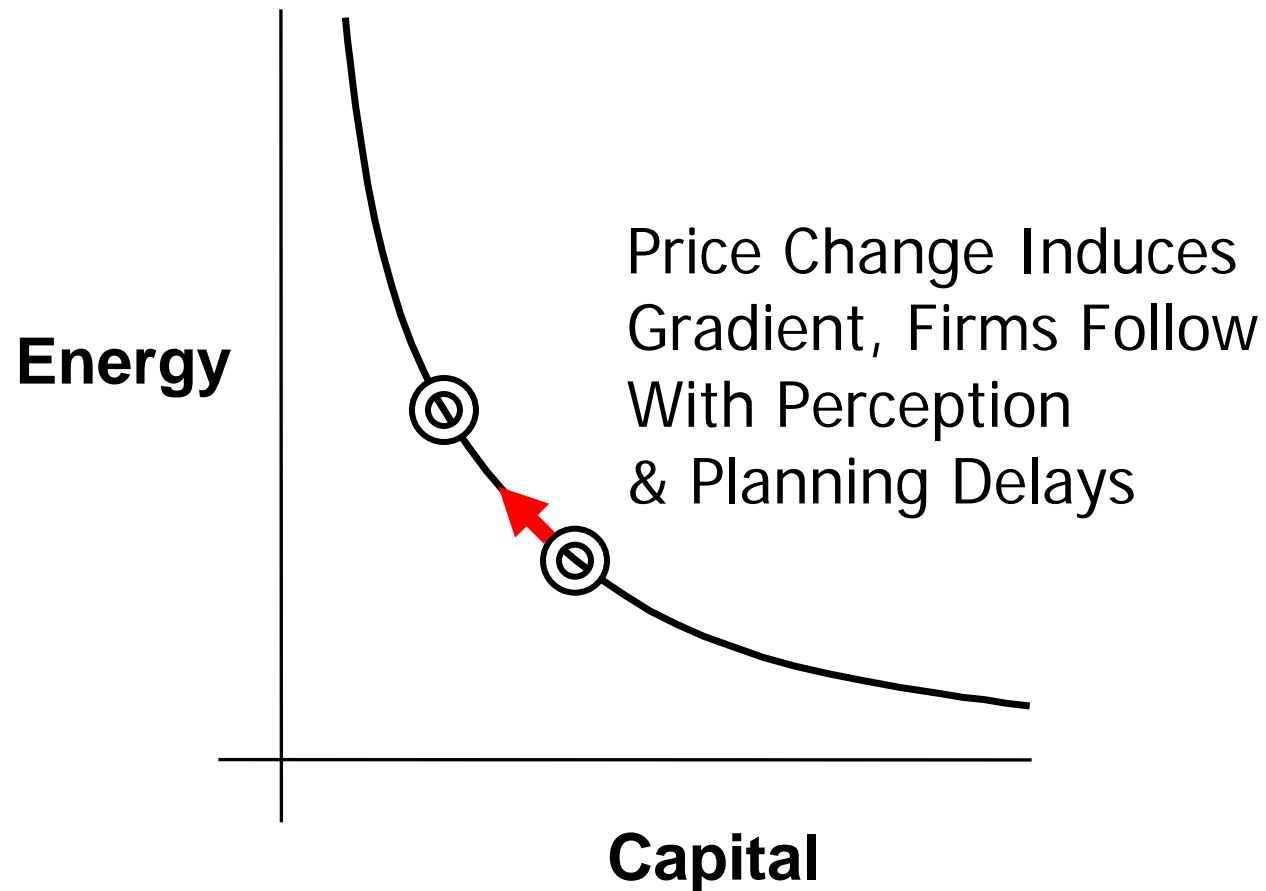
Representing Behavior



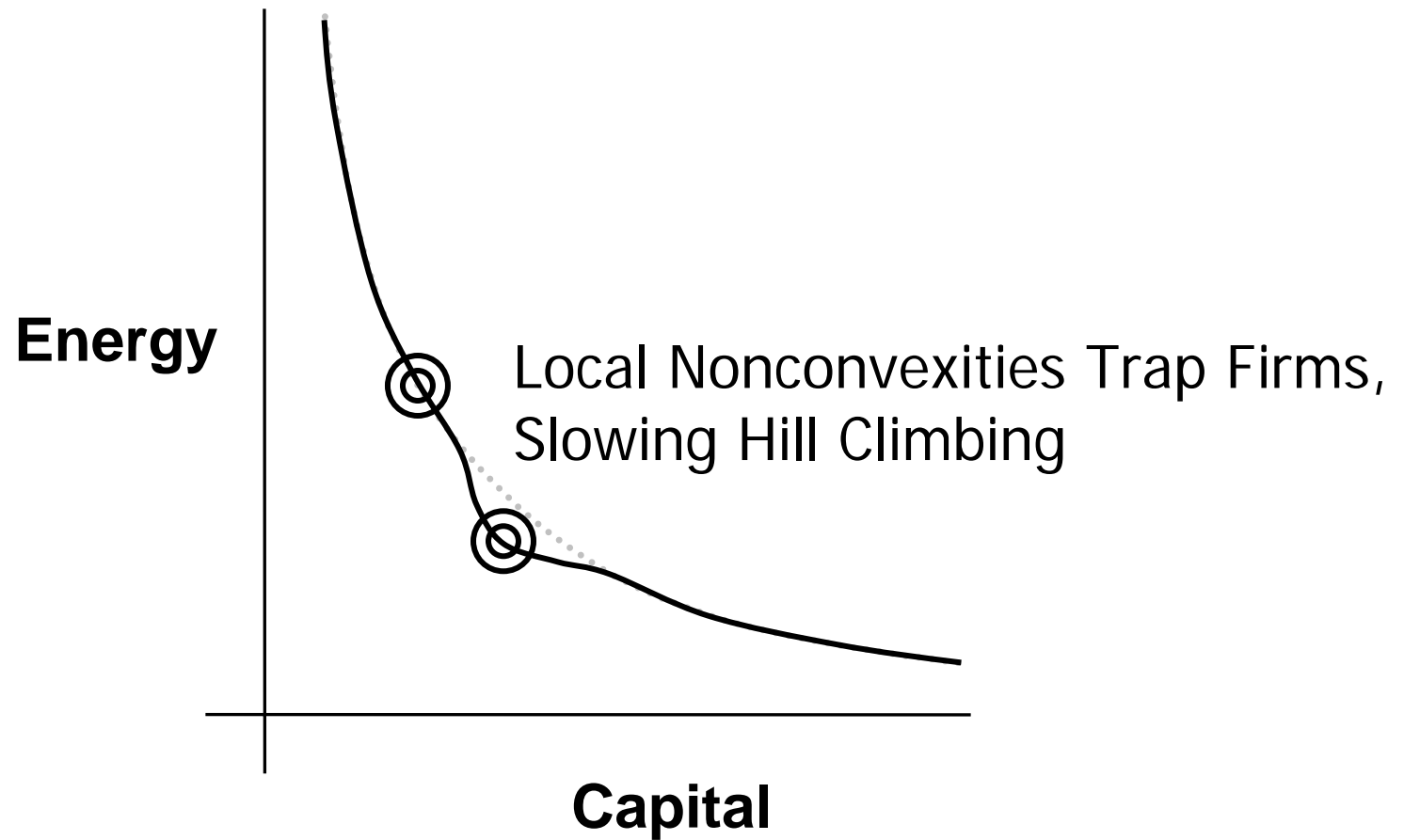
Production Technology



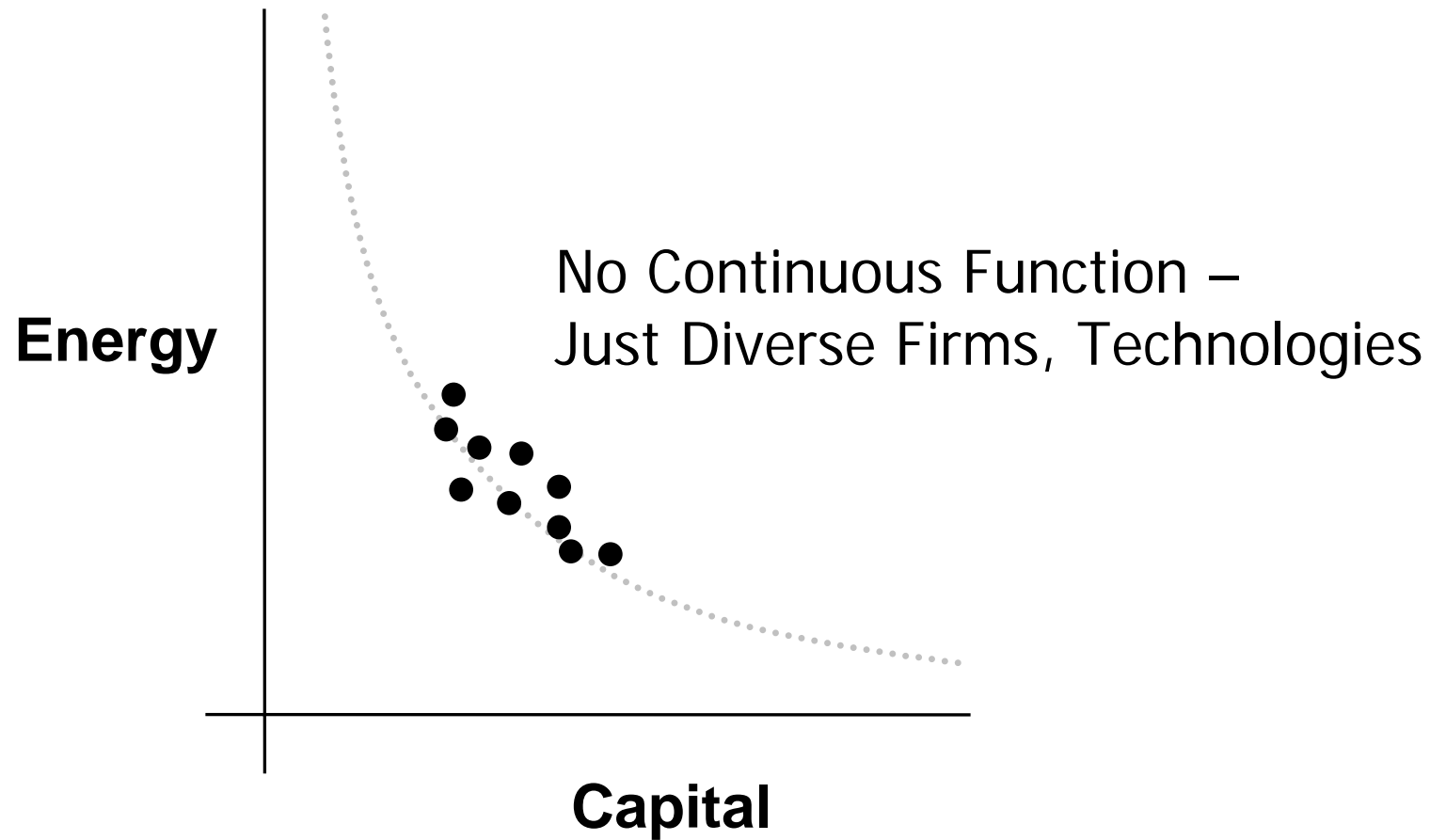
Production Technology



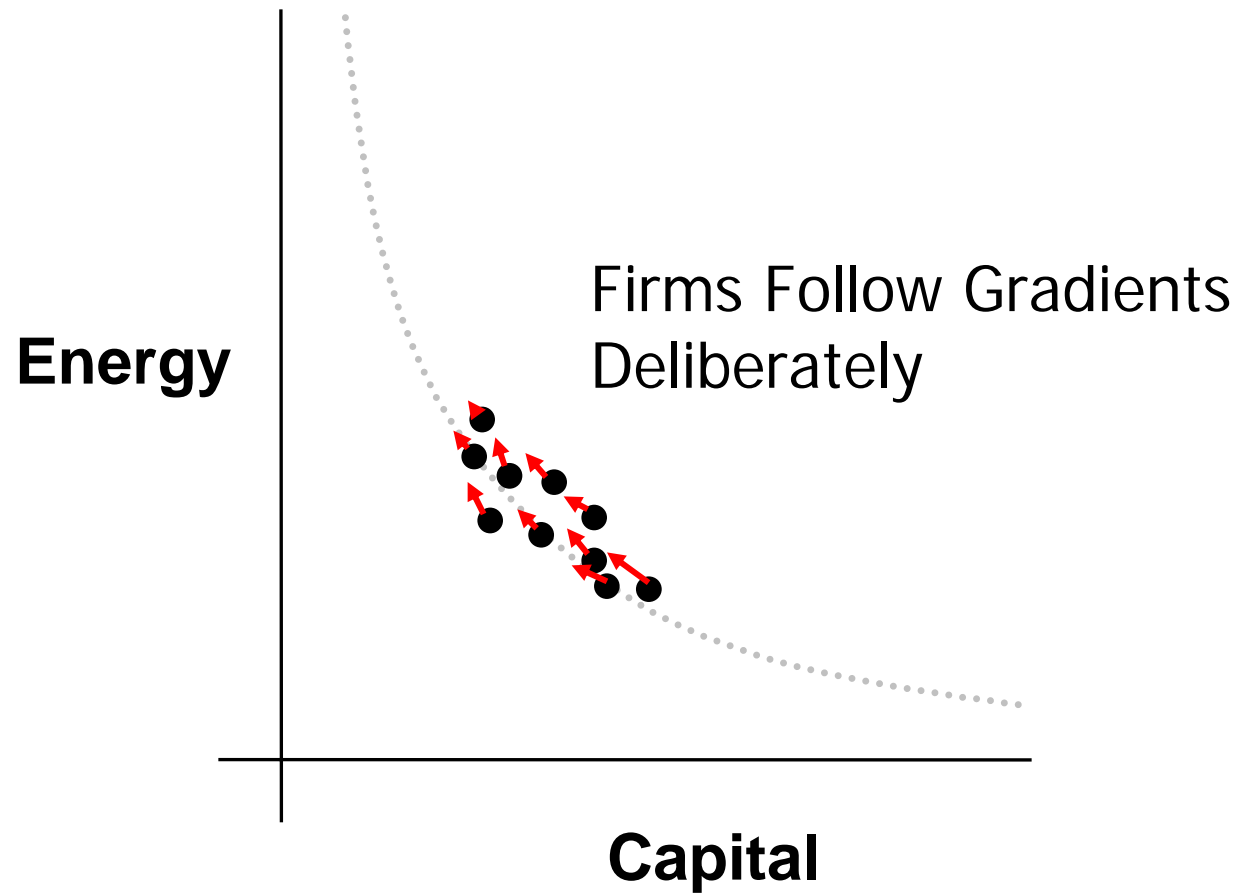
Production Technology



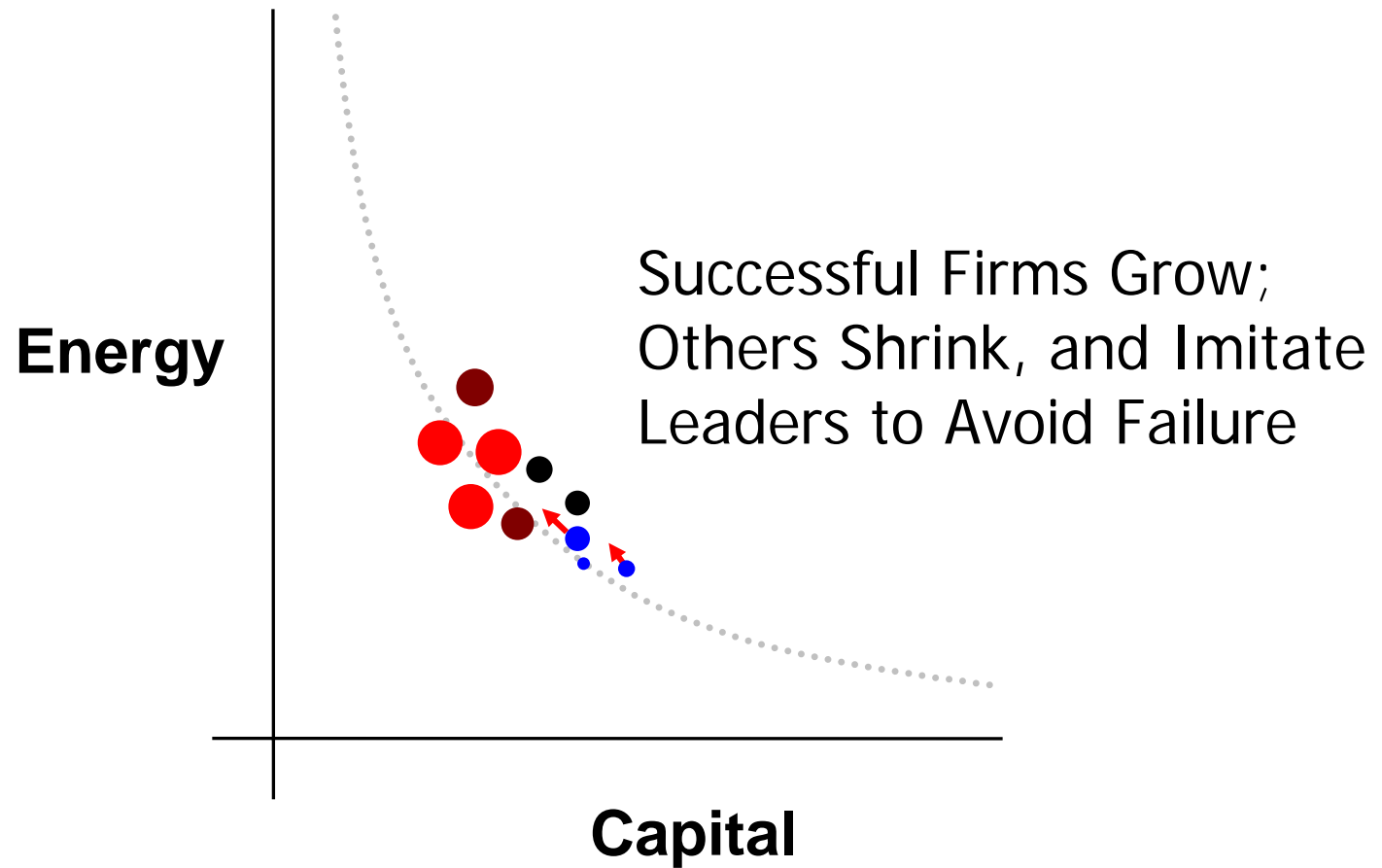
Production Technology



Production Technology



Production Technology



Micro Foundations of Macro Behavior

- At the most behavioral level of description, it's possible to see *really* bottom-up emergence of macro phenomena

But...

- Many more firm structures and parameters to specify
- Slower model execution
- Cumbersome calibration to data
- Inventories needed to buffer non-clearing markets
- Difficult to represent detail (e.g. sectors) in a recognizable way

Static vs. Dynamic Tasks

- **Factor allocation is dynamically fairly simple (obvious gradient, quick feedback on performance)**
- **Other problem domains are dynamically complex**
 - Intertemporal allocations
 - Large project management
 - Networks
 - Preferences
 - *R&D*
- **In dynamically complex environments, the system is likely to evolve faster than equilibrium can emerge**

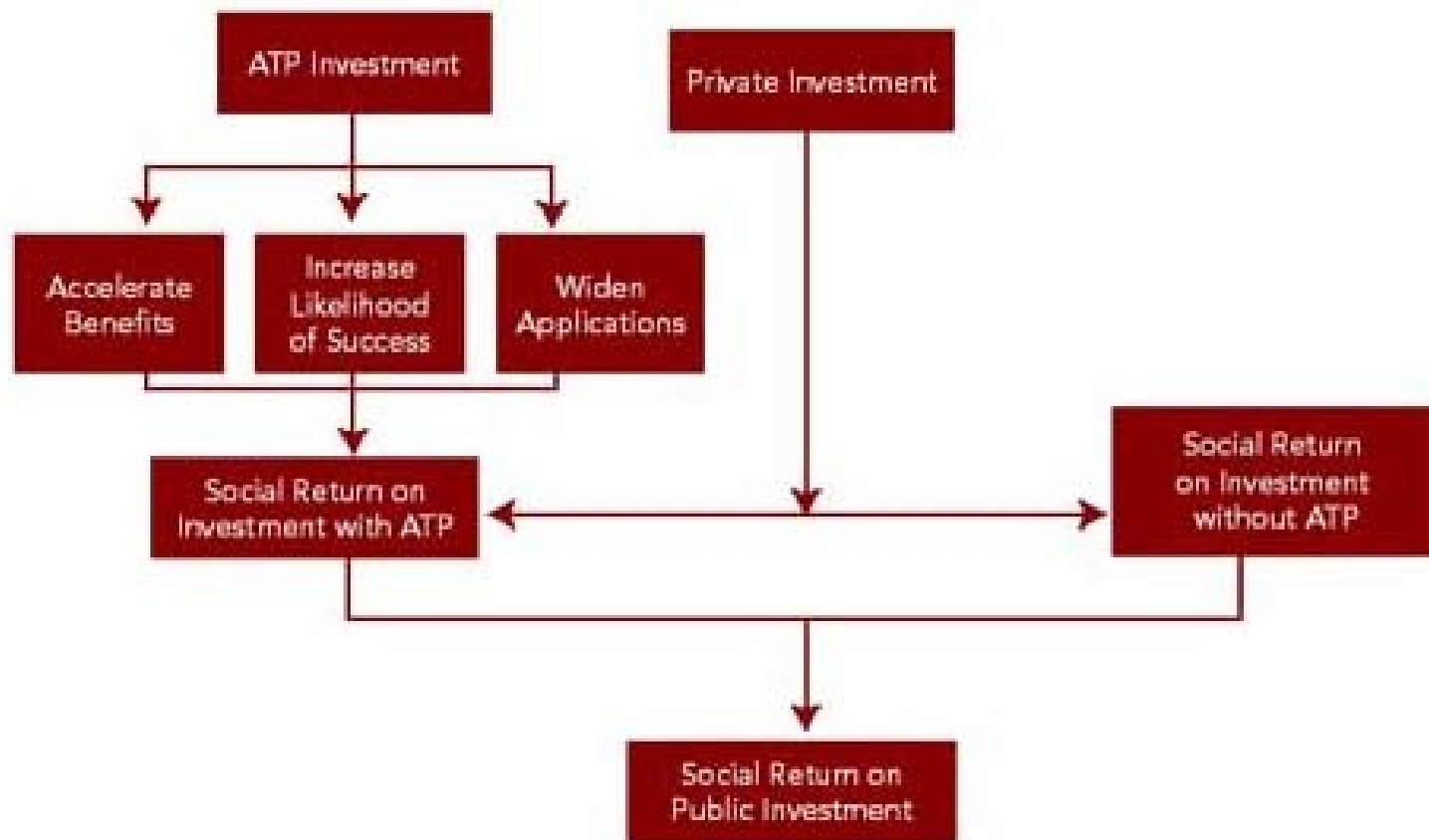
Energy Meets Science Policy: Endogenous Technology

- **Lots of progress**
 - Implementation of learning curves and deliberate R&D
- **Some progress**
 - Diffusion and adoption dynamics
 - Human capital measurement
 - Normative R&D policy
- **Limited progress on other fronts**
 - Spillovers, particularly across disciplines
 - Cannibalization and crowding of funding
 - Operational explanation of learning
 - Behavioral R&D policy

Portfolio Evaluation

- **Technologies interact in important ways**
 - Some useful outcomes (e.g. H2 economy) require multiple coordinated successes
 - Other technologies represent redundant approaches to the same problem
 - Research projects have precedence relationships and compete for limited resources
 - Knowledge spillovers cross technology stovepipes
- **Some technologies have little value until complementary policies (markets, taxes, regulations) support them**
- **Much of the value of a technology is due to provision of non-market amenities or hedges against contingencies**
- **Research program support can be dynamically adjusted as new information arrives**

Figure 6–1. Elements Determining Social Return on Public Investment and Social Return on Investment



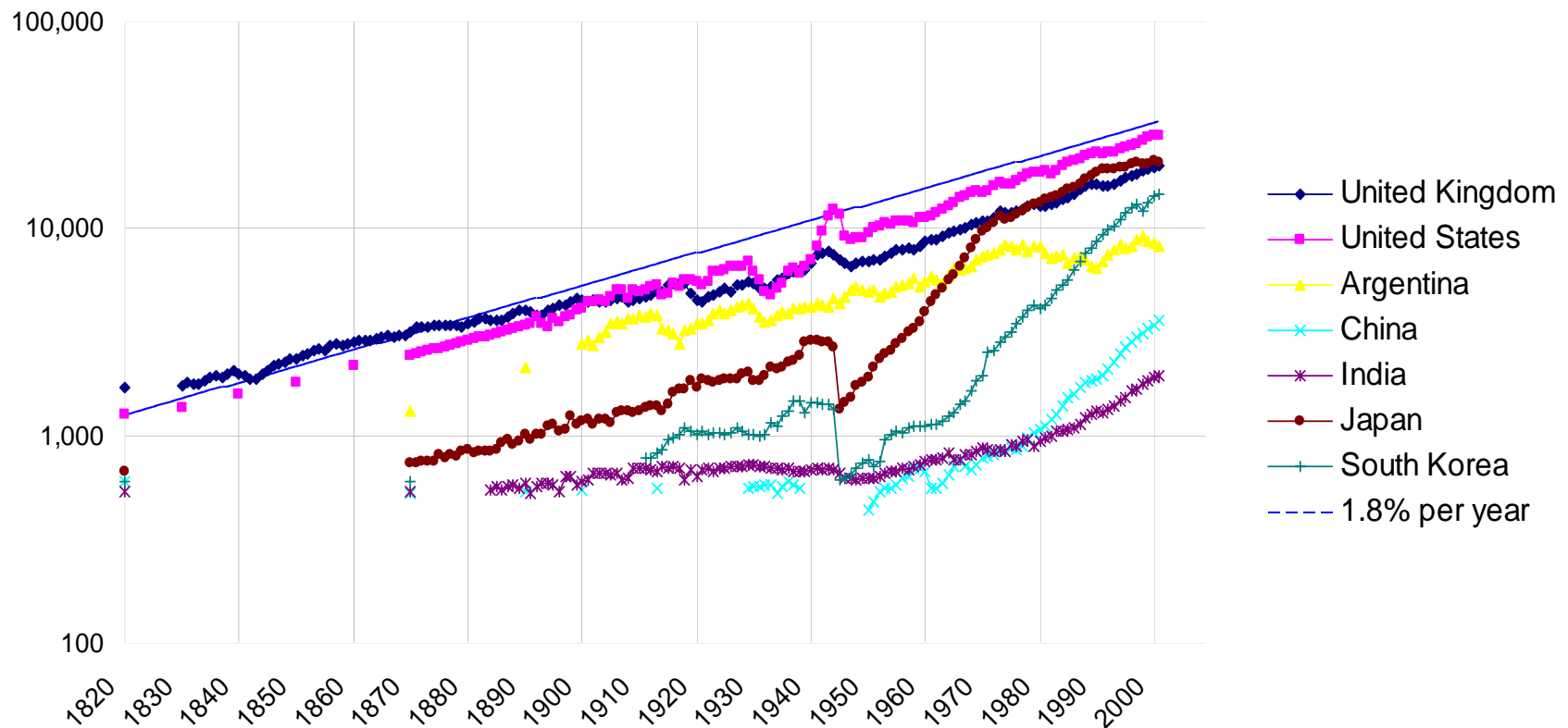
*A Toolkit for Evaluating Public R&D Investment
Models, Methods, and Findings from ATP's First Decade
Rosalie Ruegg, Irwin Feller July 2003*

The Unique Challenge in Basic Science

- **The menu is unknown**
 - What technologies will be relevant in 50 years?
 - What spillovers will current science have on the economy?
- **The appetite is unknown**
 - What strategic threats will emerge?
 - What will we want? (What is the economy for?)
- **Two approaches:**
 - Pretend we know and do lots of Monte Carlo simulation
 - Search for generic strategies that work regardless of the agenda

The economic growth of nations

Per Capita GDP
(1990 International Geary-Khamis dollars)



A Way Forward

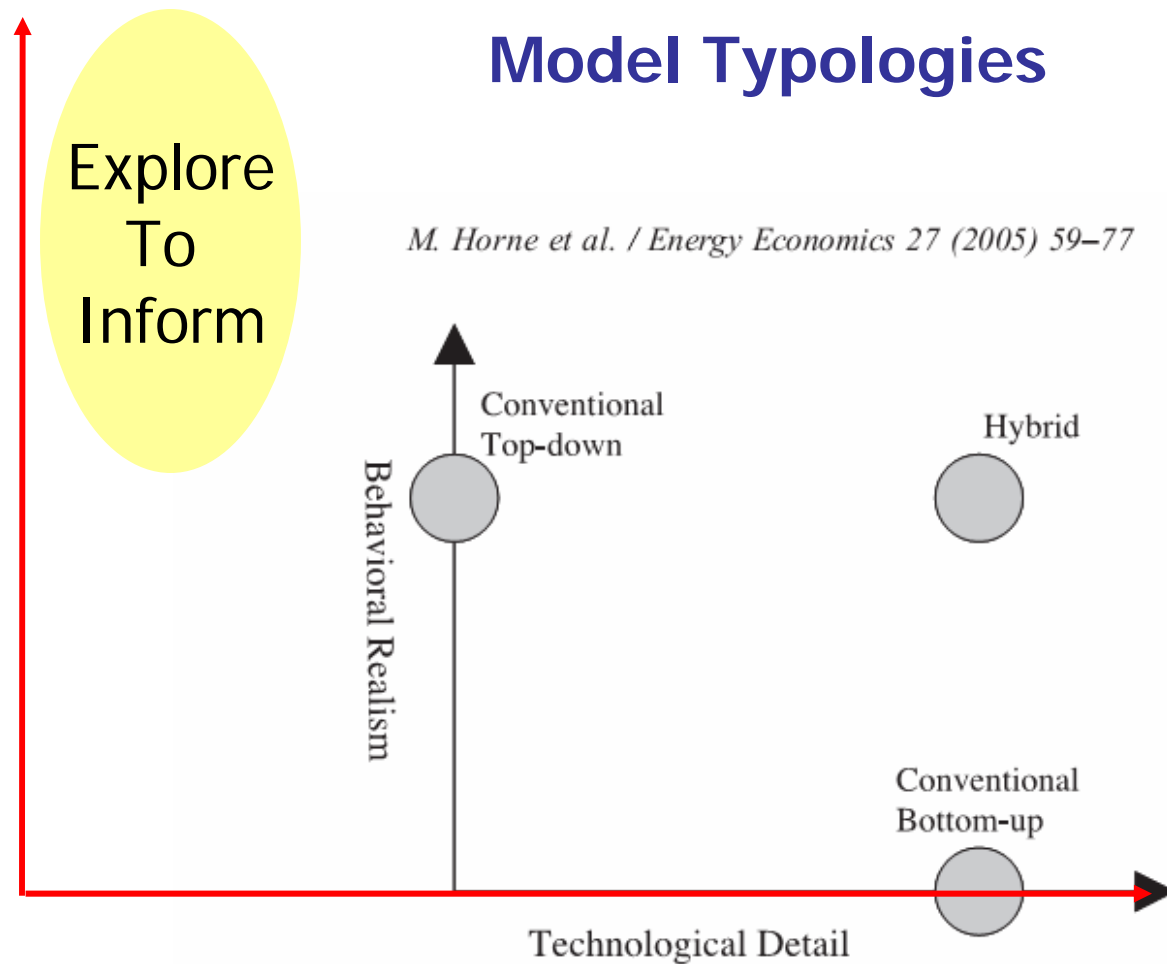
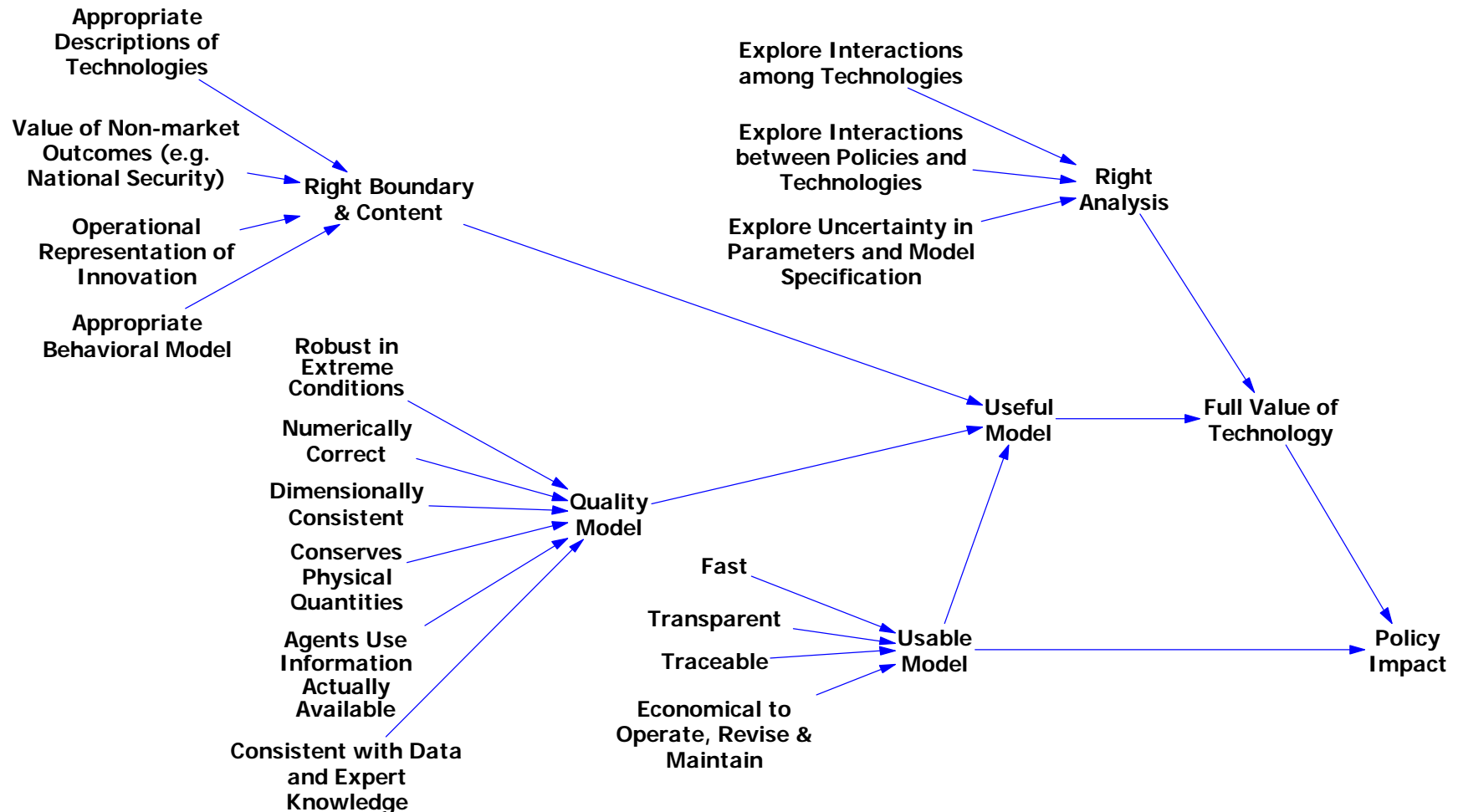


Fig. 1. Energy-economy model typologies (adopted from Jaccard et al., 2003).

A Hierarchy of Models and Questions



Balanced Mix of Critical Elements



Broad Mix of Disciplines

- **Evaluation research**
- **Psychology**
- **Org Sci**
- **Management Sci**
- **Marketing**
- **ABM/Complexity**
- **Economics**

Principles for Modeling Behavior

- **The inputs to all decision rules must be restricted to information actually available to real decision makers.**
 - Expectations about the future are based on historical information and may therefore be incorrect.
 - Actual conditions and perceived conditions differ due to measurement and reporting delays and conflicting prior beliefs.
 - The outcomes of untried contingencies are not known.
- **The decision rules of a model should conform to managerial practice.**
 - All variables and relationships should have real world counterparts and meaning.
 - Units of measure must balance without the use of arbitrary scaling factors.
 - Decision making should not be assumed to conform to any prior theory but should be investigated firsthand.
- **Desired and actual conditions should be distinguished. Physical constraints to the realization of desired outcomes must be represented.**
 - Desired and actual states should be distinguished.
 - Desired and actual rates of change should be distinguished.
- **Decision rules should be robust under extreme conditions.**
- **Equilibrium should not be assumed. Equilibrium and stability may (or may not) emerge from the interaction of the elements of the system.**

Adapted from Sterman (2000)

Three Challenges

- **Bottom-up foundations for top-down models: how to develop global model structures consistent with aggregates of populations of realistic firms?**
- **Communication: how to maintain transparency and usability when doing things right implies more structure?**
- **Productivity:**
 - Faster model building
 - Efficient federation of models
 - Use of data
 - Automation of robustness checks
 - Exploration and visualization

Examples – Behavioral Models

- **ETM (Stermann)**
- **MADIAM (Weber)**
- **FREE (Fiddaman)**
- **China Coal (Ventana for NETL)**
- **E3 (Ventana for DOE SC)**